

Def. Let $(X_t)_{t \in I}$ be a process. The maximal function $o^+(X_t)$ is $X^*(\omega) := \sup_{t \in I} |X_t(\omega)|$.

Theorem. Let $(X_n)_{n \in \{1, 2, \dots, N\}}$ be a submartingale

$$\text{Then } \forall \lambda > 0: \lambda P(X^* \geq \lambda) \leq \int_{X^* \geq \lambda} |X_N| dP$$

Proof. Observe: $|X_n|$ is a submartingale.

$$T := \begin{cases} \min_N \{n : |X_n| \geq \lambda\}, & \text{if } X^* \geq \lambda \\ N, & \text{otherwise} \end{cases} \quad \text{- stopping time.}$$

so $T \leq N$, and

$$\begin{aligned} E(|X_N|) &\geq E(|X_T|) = \int_{X^* \geq \lambda} |X_T| dP + \int_{X^* < \lambda} |X_N| dP \geq \\ &\lambda P(X^* \geq \lambda) + \int_{X^* < \lambda} |X_N| dP. \end{aligned}$$

(subtract to get Theorem \Rightarrow)

Theorem Let $p > 1$. Let $(X_t)_{t \in I}$ be a submartingale

Assume that X_t has continuous trajectories if $I = \mathbb{R}/\mathbb{R}_+$

$$\text{Then } E(X^{*p}) \leq \left(\frac{p}{p-1}\right) \int_t^p \sup_t E(|X_t|^p).$$

Proof Let us first prove for finite $I = \{1, 2, \dots, N\}$.

Let μ be the law of X^* .

Then

$$\begin{aligned} E(X^{*p}) &= \int_0^\infty \lambda^p d\mu(\lambda) = \int_0^\infty p \lambda^{p-1} P(X^* \geq \lambda) d\lambda \leq \\ &\int_0^\infty p \lambda^{p-1} \left(\frac{1}{\lambda} \int_{X^* \geq \lambda} |X_N| dP\right) d\lambda \stackrel{\text{Fubini}}{=} p E(|X_N| \int_0^{X^*} \lambda^{p-2} d\lambda) \end{aligned}$$

$$\int_0^{\infty} p \lambda^{p-1} \left(\frac{1}{\lambda} \int_0^{\lambda} |X_N| dP \right) d\lambda \stackrel{\text{Rubbin.}}{=} p E(|X_N| \int_0^{X^+} \lambda^{p-2} d\lambda) \\ = \frac{p}{p-1} E(|X_N| (X^+)^{p-1}) \stackrel{\text{Young}}{\leq} \frac{p}{p-1} E(X^{*p})^{\frac{p-1}{p}} E(|X_N|^p)^{\frac{1}{p}}$$

For discrete time: let $N \rightarrow \infty$, then

$$X^{(N)*} := \max_{1 \leq n \leq N} |X_n| \nearrow X^*, \text{ and for each } N$$

$$E((X^{(N)*})^p) \leq \left(\frac{p}{p-1}\right)^p E(|X_N|^p) \leq \left(\frac{p}{p-1}\right)^p \sup_n E(|X_n|^p)$$

So

$$E(X^{*p}) = \lim_{N \rightarrow \infty} E((X^{(N)*})^p) \leq \left(\frac{p}{p-1}\right)^p \sup_n E(|X_n|^p).$$

For continuous time, observe that

$$X^* = \sup_{t \in \mathbb{Q}} |X_t| \text{ - by continuity.}$$

And we can represent $\mathbb{Q} = \cup D_n$ - finite sets, $D_n \subset D_{n+1}$.

As before, $X^{(n)*} := \max_{t \in D_n} |X_t| \nearrow X^*$, and

$$E(X^{*p}) = \lim_{n \rightarrow \infty} E((X^{(n)*})^p) \leq \left(\frac{p}{p-1}\right)^p \sup_{t \in \mathbb{I}} E(|X_t|^p) =$$

Corollary In the assumptions of previous corollary, $P(X^* \geq \lambda) \leq \frac{\sup_{t \in \mathbb{I}} E(|X_t|^p)}{\lambda^p}$

Proof. Discrete approximation, as before.

Corollary Let $(X_t)_{t \in \mathbb{I}}$ be a submartingale (continuous for $\mathbb{I} =]R[, R_+$) and $\sup_t E(|X_t|) < \infty$. Then $\overline{\lim}_{t \rightarrow \infty} |X_t| < \infty$ a.s.

Proof. $P(\overline{\lim}_{t \rightarrow \infty} |X_t| > \frac{1}{\varepsilon}) \leq P(X^* > \frac{1}{\varepsilon}) \leq \varepsilon \sup_t E(|X_t|) =$

Thm (Martingale convergence Theorem)

Let $p > 1$, $(X_t)_{t \in I}$ - martingale (continuous in continuous time),
 $\sup_{t \in I} E(|X_t|^p) < \infty$. Then $\exists X_\infty$:

- 1) $E(|X_\infty|^p) < \infty$, X_∞ is \mathcal{F}_∞ -measurable.
- 2) $X_t \rightarrow X_\infty$ a.s. and in L^p (for $p < \infty$).
- 3) $X_t = E(X_\infty | \mathcal{F}_t)$.

Proof. Enough to prove for $p < \infty$ (since $L^\infty \subset L^p$, and a.s. limit of functions from L^∞ is in L^∞).

Take a subsequence X_{t_k} weakly (in L^p) converging to some X . X - \mathcal{F}_∞ -measurable (since so are all X_{t_k}).

Let $A \in \mathcal{F}_s$ for some s . Then $\mathbb{1}_A \in L^q$, so
$$\int_A X dP = \lim_{k \rightarrow \infty} \int_\Omega X_{t_k} \mathbb{1}_A dP = \lim_{k \rightarrow \infty} \int_A X_{t_k} dP = \int_A X_s dP,$$
 since for $t_k > s$, $E(X_{t_k} | \mathcal{F}_s) = X_s$.

So $X_s = E(X | \mathcal{F}_s)$. Since $\mathcal{F}_\infty = \sigma(\cup \mathcal{F}_s)$, X is uniquely defined
So, by compactness, $X_t \xrightarrow{w} X$.

A.s. convergence:

Let $\Omega_X(\omega) := \lim_{t \rightarrow \infty} X_t(\omega) - \lim_{t \rightarrow \infty} X_t(\omega)$. It is a.s. finite
($\lim |X_t| < \infty$ a.s.).

Also $P(\Omega_X > \varepsilon) \leq P(\lim |X_t| \geq \frac{\varepsilon}{2}) < \frac{2}{\varepsilon} \int |X_\infty| dP$.

Let now $\mathcal{D} := \{g \in L^p : g \text{ is } \mathcal{F}_s\text{-measurable for some } s\}$.

\mathcal{D} is dense in L^p .

So $\forall \varepsilon > 0 \exists g_\varepsilon : E(|X - g_\varepsilon|^p) < \varepsilon^{2p} \Rightarrow E(|X - g_\varepsilon|) < \varepsilon^2$

Observe; $\Omega_{g_\varepsilon} = 0$ (since $\lim_{t \rightarrow \infty} X_t = \lim_{t \rightarrow \infty} X_t = X_\infty$), so

$$P(\Omega_X(\omega) > \varepsilon) = P(\Omega_{X - g_\varepsilon}(\omega) > \varepsilon) \leq \frac{2}{\varepsilon} \cdot \varepsilon^2 = 2\varepsilon.$$

So $\Omega_X = 0$ a.s., and $\lim_{t \rightarrow \infty} X_t =: Y$ a.s.

L^p -convergence: $\|X_t - X\|_p \leq \|X_t - g_\varepsilon\|_p + \|g_\varepsilon - X\|_p =$

$$E(E(|X - g_\varepsilon|^p | \mathcal{F}_t))^{1/p} + \|g_\varepsilon - X\|_p \leq 2\|g_\varepsilon - X\|_p \leq 2\varepsilon^2.$$

So $\|X_t - X\|_p \xrightarrow{t \rightarrow \infty} 0$.

Why is $X = Y$? Because $\exists X_{t_n} \xrightarrow{a.s.} X \Rightarrow X = Y$

Def A family $(X_t)_{t \in I}$ is uniformly integrable if

$$1) \forall \varepsilon > 0 \exists \delta > 0 : \forall t \forall E \in \mathcal{F}_t, P(E) < \delta \Rightarrow \int_E |X_t| dP < \varepsilon. \quad g_\varepsilon$$

$$2) \sup E(|X_t|) < \infty$$

(1) \Rightarrow 2) for Lebesgue

Observe X_t - uniformly integrable, by Chebyshev inequality

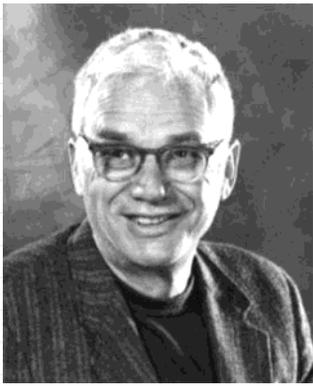
$$\int_E |X_t| dP \stackrel{\text{Young}}{\leq} \left(\int_E |X_t|^p dP \right)^{1/p} (P(E))^{1/q} \leq \delta^{1/q} \sup E(|X_t|^p)^{1/p}.$$

So, by standard convergence theory, $X_t \rightarrow Y$ a.s. $\Rightarrow \|X_t - Y\|_1 \rightarrow 0$

$$\left(\int |X_n - Y| dP \leq \varepsilon P(|X_n - Y| \leq \varepsilon) + \int_{|X_n - Y| > \varepsilon} |X_n - Y| dP \right)$$

since $\leq \varepsilon$ for large n ,
since $P(|X_n - Y| > \varepsilon) \rightarrow 0$

So, since $\|X_t - X\|_p \rightarrow 0$, $X = Y$



Joseph L. Doob (1910-2004)

Thm (Doob's martingale convergence).

Let $(X_t)_{t \in I}$ be a submartingale (continuous if $I = \mathbb{R}/\mathbb{R}_+$).

Let $\sup_{t \in I} E(|X_t|) < \infty$. Then $\exists \lim_{t \rightarrow \infty} X_t = X_\infty$ a.s.

(Enough even to assume $\sup E(X_t^+) < \infty$ ($X_t^+ = \max(X_t, 0)$))

Remark. Same is true for supermartingales (X_t) with $\sup E(X_t^-) < \infty$ (consider submartingale $(-X_t)$).

Proof relies on the following technical lemma.

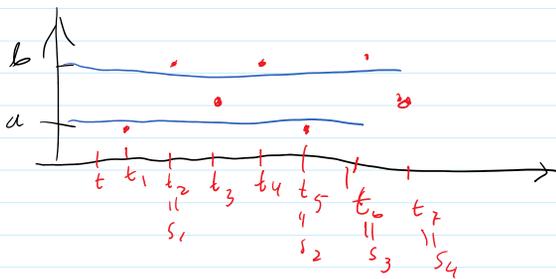
For a function $f: \mathbb{R}_+ \rightarrow \mathbb{R}_+$ (or $\mathbb{N} \rightarrow \mathbb{R}_+$),

let $\{t_1 < t_2 < \dots < t_d\} = F$ -partition.

$s_1 := \min \{t_i : f(t_i) > \beta\}$, $s_2 := \min \{t_i : s_1 < t_i < s_1 : f(t_i) < \alpha\}$.

$s_{2n+1} := \min \{t_i : s_{2n} < t_i < s_{2n} : f(t_i) > \beta\}$ $s_{2n+2} := \min \{t_i : s_{2n+1} < t_i < s_{2n+1} : f(t_i) < \alpha\}$.

(Convention: $\min \emptyset = t_d$)



$D(f, F, [a, b])$

$D(f, [a, b]) := \sup_{\tilde{F}} \max \{n : s_{2n} < t_d\}$ - number of crossings from β to α .

Lemma. Let (X_t) be a submartingale, then

$$\forall a < b, (b-a) E[D(X, [a, b])] \leq \sup_{t \in I} E[(X_t - b)^+]$$

$$((X_t - b)^+ := \max(X_t - b, 0))$$

Proof. Enough to check $\forall F$.

s_n -stopping times. $A_k := \{\omega : s_k < t_d\} \in \mathcal{F}_{s_k}$.

Moreover, $A_k \supset A_{k+1}$, $X_{s_{2n-1}} > b$ on A_{2n-1} , $X_{s_{2n}} < a$ on A_{2n} .

Therefore, by finite Optional Stopping Time Theorem:

$$0 \leq \int_{A_{2n-1}} (X_{s_{2n-1}} - b) dP \stackrel{\text{Submartingale}}{\leq} \int_{A_{2n-1}} (X_{s_{2n}} - b) dP \leq (a-b) P(A_{2n}) +$$

$$\int_{A_{2n-1} \setminus A_{2n}} (X_{s_{2n}} - b) dP. \text{ But on } A_{2n}^c, \text{ we have } s_{2n} = t_d.$$

$$\text{So } (b-a) P(A_{2n}) \leq \int_{A_{2n-1} \setminus A_{2n}} (X_{t_d} - b)^+ dP.$$

Now observe:

$A_{2n} = \{D(X, [a, b]) > n\}$, $A_{2n} \setminus A_{2n-1}$ - disjoint for different n . So just sum up! \cong

Proof of Doob.

Assume that X_t is not a.s. convergent as $t \rightarrow \infty$

Then $\exists a < b : \underline{\lim} X_t < a < b < \overline{\lim} X_t$ with positive probability. So $D(X, [a, b]) = \infty$ with positive probability:

contradiction with the Lemma! \cong

We are ready to answer the question:

When is (X_t) generated by some X_∞ , i.e.

$$X_t = E(X_\infty | \mathcal{F}_t)!$$

Thm $\text{TI} \iff A$ equivalent for a martingale $(X_t)_{t \in I}$ (assumed to be continuous when $I = \mathbb{R} / \mathbb{R}_+$).

$$1) \exists \lim_{t \rightarrow \infty} X_t \text{ in } L^1.$$

$$2) \exists X_\infty : E(X_\infty | \mathcal{F}_t) = X_t.$$

3) X_t is uniformly integrable.

Proof. 2) \Rightarrow 3) obvious

$$1) \Rightarrow 2) \quad E(X_\infty | \mathcal{F}_t) \stackrel{L^1}{=} \lim_{s \rightarrow \infty} E(X_{t+s} | \mathcal{F}_t) = X_t.$$

$$3) \Rightarrow 1) \quad \sup_{t \in \mathbb{I}} E(|X_t|) < \infty \stackrel{\text{Doob}}{\Rightarrow} \exists \lim_{t \rightarrow \infty} X_t =: X_\infty \text{ a.s.}$$

By uniform integrability, $X_t \rightarrow X_\infty$ in L^1 . \equiv

Thm (General optional stopping time theorem)

Let (X_t) be a martingale, $S \leq T$ -stopping times

$$1) \text{ If } T \text{ is bounded then } X_S = E(X_T | \mathcal{F}_S).$$

2) If X is uniformly integrable, then

$$X_S = E(X_T | \mathcal{F}_S) = E(X_\infty | \mathcal{F}_S).$$

Reminder. $\mathcal{F}_S := \{A : A \cap \{S \leq t\} \in \mathcal{F}_t \forall t\}$.

Proof. 2) \Rightarrow 1), since $(X_t)_{t \leq M}$ is uniformly integrable, $X_t = E(X_M | \mathcal{F}_t)$.

To prove 2), enough to prove

$$(*) \quad X_S = E(X_\infty | \mathcal{F}_S), \text{ since then}$$

$$X_S = E(E(X_\infty | \mathcal{F}_T) | \mathcal{F}_S) = E(X_T | \mathcal{F}_S).$$

Observe that the family $(E(X_\infty | \mathcal{B}))_{\mathcal{B} \text{-sub-}\sigma\text{-algebra of } \mathcal{F}_\infty}$ is uniformly integrable.

We know (*) for S_n -bounded finitely valued stopping times. Moreover, if S_n has finitely many values $I = \{t_1, t_2, \dots, t_n, \infty\}$, then

$$X_{S_n} = E(X_\infty | \mathcal{F}_{S_n}) \text{ (this is just a finite-indexed martingale } (X_t)_{t \in I} \text{).}$$

Let S_n -finitely valued stopping times, $S_n \downarrow S$.

Then $\mathcal{F}_S \subset \mathcal{F}_{S_n}$. (Indeed, $A \cap \{S \leq t\} \in \mathcal{F}_t \Rightarrow A \cap \{S_n \leq t\} = \underbrace{A \cap \{S \leq t\}}_{\in \mathcal{F}_t} \cap \underbrace{\{S_n \leq t\}}_{\in \mathcal{F}_t}$).

So $\forall A \in \mathcal{F}_S \Rightarrow A \in \mathcal{F}_{S_n}$, so

$$\int_A X_{S_n} dP = \int_A X_\infty dP \quad (X_{S_n} = E(X_\infty | \mathcal{F}_{S_n})).$$

$X_{S_n} \rightarrow X_S$, uniformly integrable \Rightarrow

$$\forall A \in \mathcal{F}_S \quad \int_A X_S dP = \lim_{n \rightarrow \infty} \int_A X_{S_n} dP = \int_A X_\infty dP.$$

$$\text{So } X_S = E(X_\infty | \mathcal{F}_S) \quad \#$$

Non-example. $X_t = \exp(\beta t - \frac{t}{2})$ - positive martingale.

$$\beta t - \frac{t}{2} \rightarrow -\infty \text{ a.s.} \Rightarrow X_t \rightarrow 0 \text{ a.s.}$$

Take $\tau = \inf \{ t \geq 0, X_t \leq d \}$. Then $\tau < \infty$ a.s. ($d < 1$)

$$E(X_\tau) = d \neq E(X_0) = 1.$$

Def Stopped martingale.

Let (X_t) - martingale, T - stopping time.

$$X_t^T := \begin{cases} X_t, & t \leq T \\ X_T, & t \geq T. \end{cases} \quad \text{with respect to } \overline{\mathcal{F}}_{t \wedge T} := \mathcal{F}_t \wedge \mathcal{F}_T$$

Observe: (X_t^T) is a martingale. If (X_t) was uniformly integrable, so is (X_t^T) .

Proof. If (X_t) is uniformly integrable then X_s^T .
 $S := \min(t, T)$ - stopping time, $S \leq T$, so, by OST, $\overline{\mathcal{F}}_S = \overline{\mathcal{F}}_{t \wedge T}$.
 $X_t^T = X_S = E(X_T | \overline{\mathcal{F}}_{t \wedge T})$
so X_t^T is uniformly integrable.

If (X_t) is any martingale, then
for fixed s , $(X_t)_{t \leq s}$ is uniformly integrable
so $X_t^T = E(X_s^T | \overline{\mathcal{F}}_t)$ if $t < s$, by previous ~~s~~